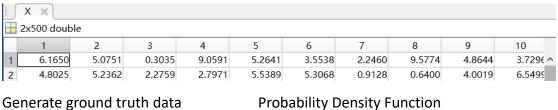
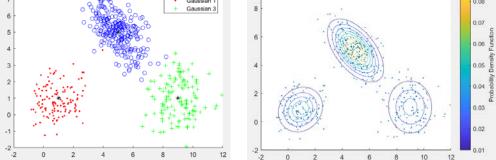
#### **EX1:**

# Use EM algorithms and the generated samples to estimate the unknown parameters

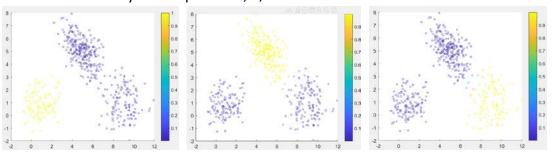
Generate 500 samples according "the first two samples are generated from the 2nd Gaussian, the 3rd sample from the 1st one, and the 4th sample from the last Gaussian".







Posterior Probability of Component 1, 2, 3



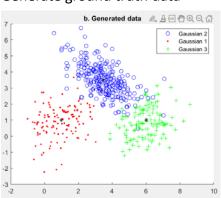
### Result:

#### b. Same as above

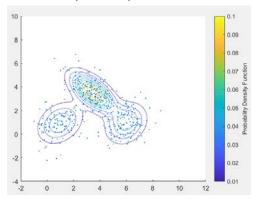
Generate 500 samples according "the first two samples are generated from the 2nd Gaussian, the 3rd sample from the 1st one, and the 4th sample from the last Gaussian".

	X X												
	2x500 double	е											
	1	2	3	4	5	6	7	8	9	10	11	12	13
1	4.0377	1.2412	1.3188	5.5664	7.0784	2.1501	1.7254	6.7147	3.3759	4.9090	1.6715	6.7172	3.98 ^
2	4.6445	5.5450	-0.0710	1.3426	3.5685	6.7379	1.2324	0.7950	4.7662	3.7883	0.1619	2.6302	4.0

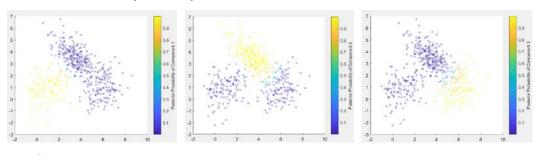
# Generate ground truth data



## **Probability Density Function**



Posterior Probability of Component 1, 2, 3

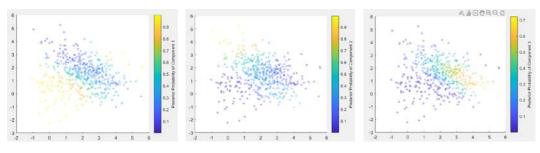


### Result:

#### c. Same as above

Generate 500 samples according "the first two samples are generated from the 2nd Gaussian, the 3rd sample from the 1st one, and the 4th sample from the last Gaussian"

_	< ×												
<u></u> 2:	x500 double	<del>)</del>											
	1	2	3	4	5	6	7	8	9	10	11	12	13
1	2.5377	-0.2588	1.3188	2.5664	5.5784	0.6501	1.7254	3.7147	1.8759	3.4090	1.6715	3.7172	2.41
2	3.1445	4.0450	-0.0710	1.3426	2.0685	5.2379	1.2324	0.7950	3.2662	2.2883	0.1619	2.6302	2.5
Ge	nerate	groun	d truth	data		Pr	obabili	ity Den	sity Fu	nction			
6.		c. G	enerated data	a		6 [						0.14	
				0	Gaussian 2	2.5		(4)					
5	0			1:	Gaussian 1 Gaussian 3	5	*	*				0.12	
4 -	89	တ ၀	o <sup>®</sup> ° °	_	Gaussian o	4 -							
3 -	8	`%& <b>\</b>	<b>1</b>	. + +		3	<u>:</u>	1	· ·			0.1 On Incipor	
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0 -	14.	‡	++++++ <del>-</del>	+ <b>6</b> 0+0 ++++	+	0 -	: 4					0.06 gg	
-1		, Elivi	+ + + +	± + + + +		-1							
-2	•					-2	149	. *				0.04	
-3	-1	0 1	2 3	. 4	5 6	-3-2	0	2	4	6	8	0.02	



Result:

In case of C. It's failed to converge in 100 iterations, which I have setting, so got the worse result.

#### d.

## (2D-visiualize as above.)

The original data set, which consists of three overlapping groups of points.

Trying to use Gaussian mixtures to get a more obvious clustering. Each one of the obtained clusters contains a significant percentage of points from more than one distribution.

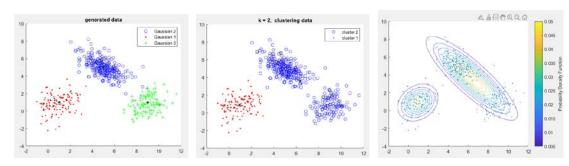
### **EX2 K-means for**

## (a) k=2

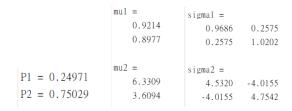
Generate ground truth data

k=2 clustering data

**Probability Density Function** 

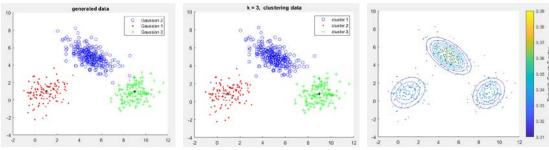


#### Result:

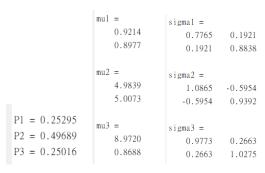


## (b) k=3





## Result:

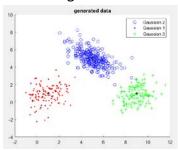


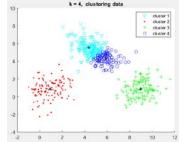
# (c) k=4

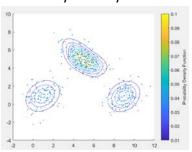
## Generate ground truth data

# k=4 clustering data

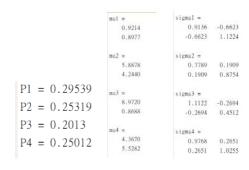
# **Probability Density Function**







### Result:

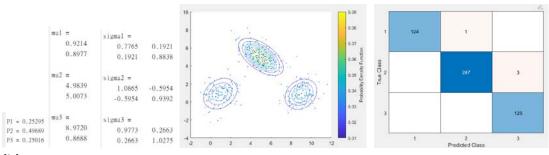


# EX3 Fuzzy k-means algorithm (k=3)

## (a) q=2

	U ×												
⊞ 3	8x500 double	9											
	1	2	3	4	5	6	7	8	9	10	11	12	13
1	0.0304	0.1613	0.9536	0.0052	0.0899	0.1447	0.9550	0.0078	0.0318	0.0402	0.9473	0.0374	0.0
2	0.9312	0.7514	0.0279	0.0114	0.5141	0.7465	0.0305	0.0149	0.9367	0.8815	0.0321	0.1052	0.97
3	0.0383	0.0872	0.0185	0.9834	0.3960	0.1088	0.0145	0.9773	0.0314	0.0783	0.0207	0.8574	0.0

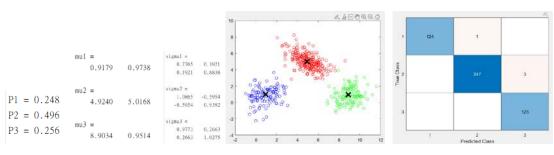
## Confusion matrix:



## (b)q=3

	U X													
	<u></u> 3x500 double													
	1	2	3	4	5	6	7	8	9	10	11	12	13	
1	0.1330	0.2591	0.7543	0.0578	0.1825	0.2436	0.7757	0.0768	0.1374	0.1424	0.7453	0.1338	0.0! ^	
2	0.7181	0.5513	0.1354	0.0855	0.4337	0.5459	0.1327	0.1058	0.7264	0.6588	0.1413	0.2238	0.79	
3	0.1490	0.1896	0.1103	0.8567	0.3838	0.2105	0.0916	0.8175	0.1362	0.1988	0.1135	0.6423	0.10	

## Confusion matrix:



FCM 的群心是藉由"所有的數據點"以"距離乘上隸屬值"而成。其中 q 值稱 為 fuzzier ,就是 FCM 中的 Fuzzy ,此值可以調控該項對於群心計算的影響 力。